

1 **Spatial associations between COVID-19 incidence rates and work sectors:**  
2 **Geospatial modelling of infection patterns among migrants in Oman**

3  
4 **\*Shawky Mansour<sup>1&2</sup> Ammar Abdulibdeh<sup>3</sup> Mohammed Alahmadi<sup>4</sup> Adham Al-Said<sup>5</sup> Alkhattab Al-**  
5 **Said<sup>5</sup> Gary Watmough<sup>6</sup> Peter M. Atkinson<sup>7&8</sup>**

6 <sup>1</sup>Geography Department, College of Arts and Social Sciences, Sultan Qaboos University, El Khodh,  
7 Aseeb, Muscat, Oman. shmansour@squ.edu.om

8 <sup>2</sup>Department of Geography and GIS, Faculty of Arts, Alexandria University, Al Shatby, P.O. Box 21526,  
9 Alexandria, Egypt.

10 <sup>3</sup>Department of Humanities, College of Arts and Sciences, Qatar University, Doha, Qatar.

11 <sup>4</sup>National Center for Remote Sensing Technology, Space and Aeronautics Research Institute, King  
12 Abdulaziz City for Science and Technology, P.O. Box 6086, Riyadh 11442, Saudi Arabia

13 <sup>5</sup>Department of Economic and Finance, College of Economics and Political Sciences, Sultan Qaboos  
14 University, El Khodh, Aseeb, Muscat, Oman.

15 <sup>6</sup>School of GeoSciences, University of Edinburgh, Surgeon's Square, Drummond Street, Edinburgh, EH8  
16 9XP, UK

17 <sup>7</sup>Lancaster Environment Centre, Lancaster University, Bailrigg, Lancaster LA1 4YR, UK

18 <sup>8</sup>Institute of Geographic Sciences and Natural Resources Research, Chinese Academy of Sciences, 11A  
19 Datun Road, Beijing 100101, China

20  
21 **\*Corresponding Author**

22 Dr. Shawky Mansour  
23 Associate Professor of GIS  
24 Department of Geography  
25 College of Arts and Social Sciences  
26 Sultan Qaboos university  
27 P.O. Box: 42 Al-Khod P.C. 123 Muscat  
28 Sultanate of Oman  
29 shmansour@squ.edu.om  
30 Tel (Office): (+ 968) 241442047  
31 Mobile: (+ 968) 96034645  
32 Fax: (+ 968) 24141851

49 **Spatial associations between COVID-19 incidence rates and work sectors:**  
50 **Geospatial modelling of infection patterns among migrants in Oman**  
51

52 **Abstract**

53 Migrants are one of the most vulnerable groups to infection with viruses due to the  
54 social and economic conditions in which they live. Therefore, spatial modelling of  
55 transmission of viruses among migrants is important for controlling and containing the  
56 COVID-19. This research focused on modelling spatial associations between COVID-  
57 19 incidence rates and migrant workers. The aim was to understand the spatial  
58 relationships between COVID-19 infection rates of migrants and the type of workplace  
59 at the sub-national level in Oman. Utilizing empirical Bayes smoothing (EBS) as well  
60 as **the** local indicator of spatial associations (LISA), six work sectors (health,  
61 agriculture, retail & business, administrative sector, manufacturing, and mining) were  
62 investigated as risk factors for disease incidence. The results indicated that the six work  
63 types each had a significant spatial association with cases of COVID-19. High rates of  
64 COVID-19 cases in relation to the workplace were clustered in the densely populated  
65 areas of Muscat. Similarly, high rates of COVID-19 cases were located in the north part  
66 of the country, along the Al-Batnah plain, where migrants are often employed in the  
67 agricultural sector. Further, COVID-19 migrants employed in the health sector was  
68 larger than **for** the other sectors. Therefore, working in the health sector can be  
69 considered a hotspot for the spread of COVID-19 infections. Due to a paucity of studies  
70 addressing the spatial analysis of COVID-19 associations with workplaces, the findings  
71 of this research are useful for decision-makers to set the necessary policies and plans to  
72 control the outbreak of the virus not only in Oman or the Gulf Cooperation Council  
73 (GCC) but also in other developing societies.

74

75 **Key words:** GIS; Spatial Associations, COVID-19 Incidence; Migrants, Work Sectors;  
76 Oman

77

78

79

80

81

82

## 83 1. Introduction

84 Efforts to control and alleviate the propagation of communicable diseases have been  
85 made on national, regional, and global scales (Chakraborty and Maity, 2020;  
86 Abulibdeh, 2020). However, efforts to reduce propagation of SARS-COV-2/COVID-  
87 19 have been hampered by several challenges, mainly the rapid transmission rate of the  
88 disease the nature of the virus associated with different socioeconomic and  
89 environmental variables (Shereen et al., 2020; Nicola et al., 2020; Gupta et al., 2020;  
90 Sohrabi et al., 2020; Li et al., 2020; Bourgonje et al., 2020). Studies have been  
91 conducted to understand the behavior of the virus, its detection, transmission,  
92 socioeconomic impacts, and treatment (Alzamora et al., 2020; Elmousalami and  
93 Hassanien 2020; Liu et al., 2020). For example, Zhai et al. (2021) utilised county-level  
94 data and geographically weighted panel regression to examine the spatiotemporal  
95 patterns of COVID-19 transmission across the U.S. There was clear spatial variation in  
96 the relation between COVID-19 transmission and socioeconomic and demographic  
97 factors. Moreover, population groups characterised by high incomes were more  
98 consistently associated low transmission rates. Similarly, Oyedotun & Moonsammy  
99 (2021) applied discrete and continuous panel regression and spatial lag models to  
100 examine the relationship between COVID-19 mortality and confirmed cases across the  
101 countries of South America. The modelling outcomes revealed that there was positive  
102 relationship between confirmed cases and population density, while there was a  
103 negative relationship between mortality and gross domestic product (GDP).

104  
105 In their recent work, Budnitz & Tranos (2021) utilised data on individual broadband  
106 speed to investigate how the quality of internet services might influence the  
107 accomplishment of work from home during the COVID-19 lockdown in the UK. The  
108 study indicated from the temporal profile of broadband usage that seven out of nine

109 clusters experienced slower upload speeds in the morning. This pattern of slow upload  
110 speed can be attributed to extreme telecommuting, particularly video calls, during the  
111 pandemic. In an analysis of the associations between staying at home during the  
112 COVID-19 pandemic and sociodemographic variables at the U.S. census block group  
113 (CBG) level, Huang et al. (2021) found a statistically significant correlation between  
114 home-dwelling time and economic status. In addition, poor communities stayed at  
115 home less compared to wealthy and rich communities, reflecting the luxury nature of  
116 stay-at-home mandates across all geographical zones. Analysing urban mobility during  
117 the COVID-19 lockdown, Kar et al. (2021) investigated spatial variation in essential  
118 travel in Columbus, Ohio during the pandemic-induced business closure. The findings  
119 showed spatial differences among socioeconomic groups in travel patterns. During the  
120 lockdown, and according to variation in spatial accessibility, moderate and high  
121 socioeconomic status travellers made large numbers of long-distance work trips while  
122 low socioeconomic status travellers made a small number of work trips.

123

124 Although no evidence has been confirmed yet that a certain group of the population is  
125 more vulnerable to COVID-19 infection, socioeconomic circumstances, living  
126 conditions and work sector have direct impacts on infection risks. Nevertheless, spatial  
127 studies that address the associations between disease transmission and demographic  
128 structure are still quite rare.

129

130 Rather than analysing the impacts of COVID-19 restrictions and lockdowns on  
131 economic sectors generally and migrant work specifically, the aim of this research was  
132 to examine the spatial associations between incidence rates among migrants and their

133 work sectors at the subnational scale in Oman. Thus, the following research questions  
134 were answered:

- 135 - To what extent is there an association between COVID-19 incidence rates  
136 among migrants and working sectors?
- 137 - Which work sectors are spatially correlated with disease incidence rates?
- 138 - Where are areas of abnormally high incidence rates given underlying work  
139 sectors as risk factors?

## 140 **2. COVID-19 and migrant populations**

141 The incidence of COVID-19 in migrants is due to the complex interaction of  
142 socioeconomic factors, health determinants, potential genetic susceptibility, higher  
143 prevalence of underlying medical co-morbidities that lead to more severe disease,  
144 and/or barriers to accessing care (Khunti et al., 2020; Tai et al., 2020). Migrants are  
145 more at risk of exposure to COVID-19 disease due to crowded conditions both in  
146 residential dwellings and in workplaces. In some countries, overcrowd work  
147 environments and low access to protection measures may have led to increased  
148 exposure and hence migrant workers highly affected by the COVID-19 pandemic  
149 (HSU, 2020; Turcotte and Savage, 2020).

### 150 ***2.1 Working Conditions/Work Sectors and COVID-19***

151 Migrant workers are typically employed in lower skilled roles and have low-income  
152 distributions, as many of them are overrepresented in lower socioeconomic strata, and  
153 hence they are forced to accept undesirable and low-status jobs (Sönmez et al., 2020).  
154 Other migrant workers may work in key services such as health care. A recent study  
155 investigated the infection and mortality rates of healthcare workers globally  
156 (Bandyopadhyay et al., 2020) and found that higher rates of infections were mainly

157 among women and nurses while the large mortality was among men and doctors. In the  
158 same vein, Kasper (2020) reported higher prevalence of COVID-19 in health sectors.

159

160 The propagation of COVID-19 has posed a serious occupational health risk to  
161 healthcare workers because of their frequent exposure to infected individuals (Saqlain,  
162 2020; Hoe Gan, 2020). Although healthcare workers may have better infectious disease  
163 protection plans than other occupational settings, they still face increased exposure to  
164 the disease and hence have higher likelihoods of being infected by the disease (Baker  
165 et al., 2020). However, these protection measures may not always be effective in  
166 protecting healthcare workers from infection, particularly when the number of incident  
167 cases is very high and hospitals are overwhelmed due to heavy patient loads  
168 (Brousseau, 2020). In Wuhan, for example, around 4% of confirmed COVID-19  
169 incidents were in healthcare workers. This indicates that such workplaces are potential  
170 locations of transmission although the workers are trained to take measures and protect  
171 themselves from such hazards (Wu and McGoogan, 2020).

172

173 Depending on occupational characteristics, some workers are at increased risk of  
174 contracting the disease due to their direct interfacing with the public and other workers  
175 and hence are considered a nexus of disease transmission to the community. Baker et  
176 al. (2020) found that around 10% of workers in the USA were employed in occupations  
177 characterized by direct exposure to disease or infection at least once a week and 18.4%  
178 of them worked in occupations where infection or exposure to disease occurs once per  
179 month (Baker et al., 2020). These workers were employed mainly in the healthcare,  
180 protective service, construction and extraction, education, office and administrative  
181 support and community and social services occupations.

182

183 Migrants commonly face adverse work conditions with low-wage jobs and experience  
184 excess chronic stress, exacerbated by economic, political, and social inequities  
185 (Sönmez et al., 2020). Migrants can also be exposed to crowded working conditions,  
186 work with hazardous material, or work while they are sick due to social or economic  
187 pressure, which may result in the further transmission of the disease (Platt and  
188 Warwick, 2020). As a result, several cases were reported in many production plants,  
189 factories, and on farms primarily by migrant workers. Migrants have been found to be  
190 less likely to work in jobs that could be performed remotely and thus are more at risk  
191 of COVID-19 (Borjas and Cassidy, 2020; Abulibdeh, 2020).

192

193 Studies found that industrial workers were more susceptible to COVID-19 infection  
194 where underlying health problems (such as respiratory system problems and  
195 occupational diseases) were common among these workers (Quynh Nguyen et al.,  
196 2017; Pham et al., 2019; Tran et al., 2020). Furthermore, self-treatment was the most  
197 common method that these workers tended to use when having health problems (Tran  
198 et al., 2019). Tran et al. (2020) concluded that the spread of COVID-19 disease among  
199 industrial workers will cause more severe and life-threatening conditions among migrants.

200

201 Hilson et al. (2020) investigated the effects of the COVID-19 pandemic on artisanal  
202 and small-scale mining (ASM) in sub-Saharan Africa. This sector employs more than  
203 25 million people and creates significant economic opportunities indirectly. In  
204 particular, this study investigated how the communities dependent on ASM have been  
205 affected economically due to the changes made at the local, regional, and international  
206 levels because of the spread of the virus. Calvimontes et al. (2020) examined the impact

207 of the pandemic on small-scale miners in different communities in Brazil. The study  
208 highlighted the capacity of miners to self-organize and find alternative solutions to cope  
209 with the pandemic. Miners in Brazil usually belong to heterogeneous social groups, and  
210 work without formal employment contracts (Calvimontes et al., 2020; de Theije, 2020).  
211 Despite the spread of the disease and the risk of being infected, economic gain is the  
212 main reason motivating miners to work, as they need to meet their financial  
213 responsibilities and basic needs. However, miners believe that they are safer in the  
214 mining site than in town in terms of the risk of disease contagion as they do not leave  
215 the mining sites very often and thus, they believe this is similar to self-quarantine.  
216 However, de Castro et al. (2020) found a disparity between the numbers of COVID-19  
217 in some mining municipalities in Brazil but jobs in this sector remained stable  
218 throughout the pandemic compared to other sectors.

219

220 Farmworkers are another group of workers that are affected by the spread of the  
221 pandemic. For example, migrant farmworkers in the USA have closed work permits  
222 and face inequities, such as overcrowded living conditions, isolation, lack of access to  
223 field sanitation, difficulty accessing health care, lack of access to personal protective  
224 equipment, and inability to switch employers (Haley et al., 2020). These conditions  
225 further magnify the possibilities of being in an increased risk of exposure to the disease  
226 (Caxaj and Plamondon, 2020). Flocks (2020) investigated the impact of the COVID-19  
227 pandemic on H-2A agricultural workers in the USA. This group of workers is more  
228 vulnerable to transmission of the disease because they have less control over their  
229 physical environment compared with domestic farmworkers. Despite the increased  
230 number of incidence cases in this group of workers, there was no rule in the US that  
231 addressed the need to protect workers' health because of the disease.



232 ***2.2 Living Conditions of Migrants and COVID-19***

233 Most of the studies conducted in the USA and Europe investigated the effects of  
234 COVID-19 on undocumented migrants (Keller and Wagner, 2020; Wernly et al., 2020).  
235 In Europe, migrants are more likely to reside in places with high population density  
236 such as urban areas or migrant campuses (Costa and de Valk, 2018). Under these  
237 conditions, migrants are unable to follow basic mitigation and prevention practices such  
238 as hand hygiene, self-isolation, and social distancing. Therefore, the COVID-19  
239 mitigation measures that most European countries are taking are difficult to implement  
240 in these areas (Maroko et al., 2020). Migrants may live in crowded multi-generational  
241 households due to cultural and economic reasons, which may increase the risk of  
242 disease transmission within households and hence mitigation measures such as social  
243 distancing or self-isolation are difficult. In the USA and the UK, migrant groups are  
244 categorized amongst vulnerable and marginalized populations with longstanding  
245 structural inequalities and complex socioeconomic health determinants (Greenaway et  
246 al., 2020; Openshaw and Travassos, 2020). The focus of these migrants living  
247 conditions have been on those living in refugee camps, reception centers and detention  
248 centers which are considered at high risk for COVID-19 infection (Greenaway et al.,  
249 2020; European Centre for Disease Prevention and Control, 2020).

250

251 Keller and Wagner (2020) studied the impact of illegal migrants detained by  
252 Immigration Customs Enforcement (ICE) in the USA in addition to the impact on the  
253 nation's public health. The ICE system is well known for overcrowding, inadequate  
254 health care, poor sanitation, and the complexity of containing contagious disease.  
255 Therefore, implementing the COVID-19 mitigation measures in these facilities is not  
256 possible and hence the authors suggested that the ICE should release those detained

257 migrants that pose no threats to public health and safety in order to prevent the spread  
258 of the disease (Keller and Wagner 2020).

259

### 260 ***2.3 Underlying Health conditions of Migrants and COVID-19***

261 Migrants are a heterogeneous population with different cultures and various health  
262 needs and, in many countries, they face barriers to health care systems (Greenaway and  
263 Castelli, 2019). COVID-19 disease has been documented in refugee camps and among  
264 asylum seekers in many countries such as Portugal, Germany, the Greek mainland, and  
265 the UK (European Centre for Disease Prevention and Control, 2020). Migrants also  
266 may face difficulties in accessing the health care system due to poverty, racial  
267 discrimination, cultural and linguistic barriers, lack of entitlement to health care or  
268 sectors within the health system, and difficulties navigating the health care system.  
269 These migrant communities lack accessible and affordable healthcare (Hardy et al.,  
270 2012; Martinez et al., 2015), which is particularly problematic during the COVID-19  
271 pandemic. Migrants in the US are also under- or uninsured which significantly limits  
272 their ability for early COVID-19 diagnosis, which is essential to optimize individual  
273 patient outcomes and to mitigate and prevent the transmission of the disease to other  
274 community members (Clark et al., 2020).

275

276 Wernly et al., (2020) investigated the cardiovascular disease (CVD) risk distribution  
277 and health literacy of migrants in Europe compared to the host population. They found  
278 that health literacy among migrants was lower than the host population and that the  
279 rates of CVD risk factors in some subgroups of migrants were higher. Furthermore,  
280 they found that health literacy and CVD risk factors were key components in managing  
281 the public health response during the COVID-19 pandemic. They concluded that

282 migrants are both a challenge and opportunity in terms of the health care system.  
283 However, interventions to manage chronic disease, screen for risk factors, and increase  
284 health literacy among migrants could improve the long-term health care outcomes for  
285 them.

286

287 Several studies have investigated the links between COVID-19 infections and living  
288 conditions and the underlying health conditions of migrants. However, few have looked  
289 at the spatial relationships between COVID-19 infection rates and work sector in  
290 migrant workers compared to the wider population. Many studies in the USA and  
291 Europe have revealed major health inequities among migrants, ethnic and racial  
292 minorities. Social and economic determinants will strongly influence susceptibility to  
293 and health outcomes of COVID-19; therefore, it is crucial to understand these factors  
294 when investigating the effect of the COVID-19 pandemic on migrants in Oman and if  
295 the migrants are disproportionately affected by the pandemic. Spatial modeling of  
296 disease transmission among migrants is crucial to guide responses to the pandemic in  
297 practice. In this study, infection rates of the disease were correlated with employment  
298 sector using spatial models to investigate the links for six work sectors within Oman,  
299 a country that has a substantial migrant population.

300

### 301 **3. Materials and methods**

#### 302 ***3.1 Study area***

303 Oman is located in the far south-eastern corner of the Arabian Peninsula, with an area  
304 of 309,501 km<sup>2</sup> and is the second largest country after Saudi Arabia on the peninsula.  
305 The country extends from latitudes 16.40°–26.20° north and longitudes 51.50°–59.40°  
306 east. Oman is bordered on the north by the Strait of Hormuz, to the northeast by the Sea

307 of Oman. On the east and south by the Arabian Sea while on the southwest by the  
308 Republic of Yemen. Oman also shares borders with the United Arab Emirates (UAE)  
309 in the north and west, and the Kingdom of Saudi Arabia (Empty Quarter) in the west  
310 (Fig.1).

311 In 2021, the total population was 4,507,323 and approximately, 61.10% (2,751,842) of  
312 them were Omani with migrant populations accounting for 1,755,481 (38.9%)  
313 (NCSI,2021). Administratively, the country is divided into 11 governorates and 61  
314 Wilayats (states), with the Muscat governorate being the largest in terms of population  
315 size and comprising six Wilayats including Muscat City as the capital of Oman. In  
316 2019, the largest population density was 355.4 person/km<sup>2</sup> in Muscat while the lowest  
317 density was in Al-Wusta governorate (0.6). The majority of migrants are concentrated  
318 in Muscat governorate (more than 850,000), while approximately 260,000 live in north  
319 Al-Batnah and slightly more than 200,000 population live in Dhofar governorate (Year  
320 book, 2020).

321 Oman is a capital rich and labor poor country, which is reflected in the high prevalence  
322 of migrant laborers. Currently, around 39% of the total population living in Oman are  
323 migrants (World Bank 2021), the majority of whom are from South Asia (India,  
324 Pakistan, Sri Lanka, Bangladesh) and Southeast Asia (Indonesia and Philippines). In  
325 Oman, migrants are pillars to several urban and rural economic sectors such as the  
326 agricultural, industrial and commercial sectors (Mansour, 2017). As a host country,  
327 Oman benefits economically from migrant labours in different aspects such as reducing  
328 wages, small businesses, and fostering entrepreneurial activities, particularly in  
329 unskilled jobs (Mansour, 2017). Oman relies heavily on importing foreign labourers  
330 from different countries mainly because of the small size of the Omani population.  
331 Furthermore, due to cultural perspectives, the percentage of local women entering the

332 workforce is low and hence the country relies on expatriate women, particularly in  
333 household jobs from South and Southeast Asian countries such as India, the Philippines,  
334 Sri Lanka, Indonesia, and Bangladesh (Mansour, 2017; Fernandez, 2010, 2011;  
335 Willoughby, 2006; Shah, 2004a, 2004b). Therefore, migrants are changing the  
336 population structure and the socioeconomic status and contributing to transformational  
337 development in the country.

338

339 The migrant workforce in Oman works mainly in construction, food services,  
340 manufacturing industries, healthcare and accommodation. These types of jobs are  
341 essential and often require workers to attend the workplace in-person as the jobs do not  
342 lend themselves to working from home. Furthermore, migrants are more likely to live  
343 with multiple roommates or with multigenerational family groups. Usually low-skilled  
344 migrants live in housing units that are characterised by overcrowding and low-quality  
345 infrastructure meaning that social distancing and self-isolation are difficult (David,  
346 2009).

347

348 Following the onset of the COVI-19 pandemic, Oman implemented strict measures and  
349 restrictions to lessen the transmission of the disease. However, the incidence rate has  
350 increased since the first confirmed registered case on 28<sup>th</sup> of February 2020 to reach  
351 133,728 confirmed cases with almost 26,000 cases per million as of January 2021  
352 (Worldometer, 2021 – end January 2021). Few studies have been conducted to  
353 investigate the spatial distribution of the disease in Oman. Mansour et al. (2021)  
354 investigated the spatial variation of the relationship between the infection rate and  
355 different sociodemographic factors such as population density, age structure, hospital  
356 beds, long-term illness, and nurse practitioners. The findings revealed that for people

357 aged 65 years and above, hospital beds, population density, and diabetes rates were  
358 statistically significant determinants of COVID-19 infection rate. Al-Kindi et al. (2020)  
359 performed spatiotemporal analysis to investigate the spatial patterns of the propagation  
360 of COVID-19 disease in Oman using five geospatial techniques. They assessed the  
361 spatiotemporal patterns of COVID-19, and quantified temporal differences in the rate  
362 of incidence. They found that the disease moved from the northeast of the country to  
363 the northwest and southwest and that the infection rates increased in the most populated  
364 areas.

365

366 In Oman, particular areas within the country have demonstrated high rates of COVID-  
367 19 cases (Mansour, 2021; Al-Kindi, 2020). Nonetheless, no studies conducted on the  
368 relationships between COVID-19 and migrants' employment in Oman nor GCC  
369 countries. These studies focused on the socioeconomic factors of migrants, particularly  
370 non-documented migrants. However, this literature did not consider the spatial patterns  
371 of COVID-19 among migrants. This gap is a major impediment to designing effective  
372 tailored interventions for this group of the population. Furthermore, although some  
373 studies investigated the impact of COVID-19 on migrants, no studies employed spatial  
374 techniques to assess the spatiotemporal characteristics of the spread of the disease on  
375 migrants. Such an investigation is crucial to quantify the spatiotemporal patterns of  
376 COVID-19 propagation across the country in general and among migrants in particular  
377 to determine the different demographic and socioeconomic factors that could accelerate  
378 transmission and incidence rates. Investigating the spatiotemporal incidence of  
379 COVID-19 is crucial in understanding the dynamics, occurrence and spread of the  
380 disease. As part of this, it is important to investigate the spatial impact of COVID-19  
381 disease on migrants in the country considering their socioeconomic characteristics.

382 Migrants in Oman work in several public and private sectors. Table 1 illustrates  
 383 descriptions of the calculated risk factors, the dependent variable and the source of each  
 384 variable.

385 **Figure 1 the study area and subnational administrative boundaries of Oman**

386

387 **Table 1 Descriptions of risk factors and data sources**

Variable	Description	Source
COVID-19 incidence rate among migrants	The COVID-19 cases in each Wilayat calculated by dividing the total number of cases by total number of migrants.	MOH, Oman
Migrants' workers in health sector	The percentage of migrants working in health sector in each Wilayat calculated by dividing the total workers in this sector by the total migrants' workers and multiplied by 100	NCSI, Oman
Migrants' workers in retails &business	The percentage of migrants working in retails &business in each Wilayat calculated by dividing the total workers in this sector by the total migrants' workers and multiplied by 100	NCSI, Oman
Migrants' workers in administrative sectors	The percentage of migrants working in administrative in each Wilayat calculated by dividing the total workers in this sector by the total migrants' workers and multiplied by 100	NCSI, Oman
Migrants' workers in agricultural activities	The percentage of migrants working in agricultural activities in each Wilayat calculated by dividing the total workers in this sector by the total migrants' workers and multiplied by 100	NCSI, Oman
Migrants' workers in manufacturing	The percentage of migrants working in manufacturing in each Wilayat calculated by dividing the total workers in this sector by the total migrants' workers and multiplied by 100	NCSI, Oman
Migrants' workers in mining	The percentage of migrants working in mining in each Wilayat calculated by dividing the total workers in this sector by the total migrants' workers and multiplied by 100	NCSI, Oman

MOH: *Ministry of Health in Oman.*

NCSI: *National Centre for Statistics and Information*

388

389

## 390 **3.2. Methods**

391 The excess risk of COVID-19 incidence can be defined as the number of infected  
392 people during the pandemic above and beyond what would be expected under normal  
393 circumstances. Spatially, in this research, we are interested in how the incidence rate of  
394 COVID-19 among migrants in each Wilayat varies compared to the national average.

395 The COVID-19 incidence rate ( $R_{cov_i}$ ) among migrants was computed as follows:

$$396 \quad R_{cov_i} = \frac{x_{cov_i}}{p_i} * 10,000$$

397 Where  $x_{cov_i}$  indicates the reported cases of disease incidence among migrants in  
398 Wilayat  $i$  while  $p_i$  refers to the total migrants' population in Wilayat  $i$ .

399

### 400 **3.2.1 Empirical Bayes Smoothing**

401 To stabilise COVID-19 incidence rates for Wilayats with small population sizes or low  
402 number of disease cases, Empirical Bayes Smoothing (EBS) was utilized. The main  
403 assumption was that the relative risks of population residing in Wilayat  $i$  ( $\delta_i$ ) were  
404 independently and symmetrically distributed according to a Poisson distribution:

$$405 \quad x_i/\delta_i \sim \text{Poisson}(N_i\delta_i)$$

406 Where  $x_i$  is the random variable illustrating disease count in Wilayat  $i$  while  $N_i$  refers  
407 to the expected count for the same zone. The Empirical Bayes Smoothed (EBS)  
408 relative risk of  $R_{cov_i}$  borrows the neighbouring Wilayat rates to adjust the uncertainty  
409 rates as per the expression:

410

$$411 \quad \hat{R}_{cov_i} = \phi R_{cov_i} + (1 - \phi_1)m\delta_i$$

412



413 Where  $\delta_i$  is the ratio of prior variance to the data variance, while  $m\delta_i$  is the prior mean  
 414 (weighted mean). The final EBS rates remains virtually unchanged for all Wilayats with  
 415 relatively large sizes of populations or numbers of cases (Nyadanu et al.,2019; Sankoh  
 416 et al.,2002,). Applying the spatial empirical Bayes (SEB) smoothed rate, spatial  
 417 patterns of the disease among migrants were mapped and visualized (Figure 2). Unlike  
 418 the raw rates, this empirical Bayes technique derives strength from neighbouring zones  
 419 substantially to minimize the effects of small populations at infection risks within  
 420 specific zones.

421

### 422 **3.3.2 Local Moran's $I$ and Local Indicators of Spatial Association (LISA)**

423 The Local Moran's  $I$  is a measure of spatial clustering of a geographical variable where  
 424 similar and dissimilar values cluster (Moran,1950). Giving the assumption that a linear  
 425 relationship exists, Moran's  $I$  is utilized specifically to identify geographical units  
 426 (Wilayats) and their neighbours where the deviations in the relationships among a  
 427 variable's observations are minimum or maximum. The local Moran's  $I$  for spatial  
 428 association is calculated as:

$$429 \quad I_i = x_i = \frac{x_i - \bar{X}}{m_2} \sum_j w_{ij}(x_i - \bar{X})$$

$$430 \quad m_2 = \frac{\sum_j(x_i - \bar{X})}{n}$$

431 Where  $n$  denotes the total number of observations (geographical zones)  
 432  $X_i$  illustrates the attribute value of a feature  $i$  (geographical zone or Wilayat) in the  
 433 considered variable and  $\bar{X}$  refers to the mean value of that variable.  $w_{ij}$  is a calculated  
 434 spatial weight between features  $i$  and  $j$  based on neighbouring status and  $m_2$  is a constant  
 435 for all locations.

436 The outcome value of Moran's  $I$  ranges from +1 (denoting positive autocorrelation and  
 437 clustering) to -1 (indicating negative autocorrelation and dispersion). The outcomes of  
 438 the analysis involve four categories of similar and dissimilar observations clustering.  
 439 The first category is **high-high** values which includes high values (above the variable  
 440 mean) clustering with other high values. The second group is the **low-low** values which  
 441 refers to a clustering pattern where low values (below the variable mean) cluster with  
 442 other low values. These two categories (HH and LL) signify a positive spatial  
 443 correlation. The remaining two groups comprise the **high-low** group which  
 444 demonstrates high values clustering with neighbouring low values (below the variable  
 445 mean). The last group is **low-high** values where low observations cluster with  
 446 surrounding high values. These two groups (HL and LH) show a Negative Spatial  
 447 Autocorrelation and illustrate dissimilar values (outliers) (Anselin, 1995).

448

### 449 **3.4.3 Bivariate local indicators of spatial association (BiLISA)**

450 The BiLISA is an explicit generalization of the concept of spatial association in which  
 451 the measure relates the value of a variable at a specific zone to that of a different  
 452 variable at the neighbouring zones. To measure the local relationship between the  
 453 COVID-19 incidence rates among migrants and a set of variables representing work  
 454 sectors, BiLISA was calculated as follows (Anselin, 1995):

455

$$456 \quad Bi_{1,2} = x_1^i \sum_{j=1}^n w_{ij} x_2^j$$

457

458 Where  $Bi_{1,2}$  refers to the bivariate value of the variable **1,2** at spatial unit  $i$  while  $n$   
 459 represents the total number of Wilayats.  $x_1^i$  indicates the degree of linear association

460 between the values of variable  $\mathbf{1}(x)$  at a specific neighbourhood  $i$ .  $x_2^j$  signifies the mean  
461 value of the second variable in neighbouring spatial zone  $j$  and  $w_{ij}$  is the spatial weight  
462 between the  $i$ th and  $j$ th spatial units.

463 To create bivariate maps, a spatial weight matrix was developed on the basis of first-  
464 order queen zone contiguity. Given the size and irregular shapes of Wilayats polygons,  
465 the queen matrix is suitable for capturing spatial autocorrelation and disease. Weights  
466 were assigned to all Wilayats that share common edges and corners where  $\Sigma w_{ij} = 1$  if  
467 Wilayat  $i$  and  $j$  shared a common boundary; otherwise  $\Sigma w_{ij} = 0$ , for non-neighbouring  
468 Wilayats.

#### 469 **4. Results**

470 In this research, we were interested in finding out how the incidence rate of COVID-19  
471 amongst migrant workers was related to the national average and how it correlated with  
472 job type. The spatial distribution of incidence rates (Fig. 2) shows that COVID-19  
473 incidence rates varied across the country with the largest numbers in the northern  
474 regions, particularly Muscat governorate and urban Wilayats (Muscat, Matrah,  
475 Bowsher, and Al-Seeb). This is due mainly to population concentration and  
476 overcrowding of migrants' housing units, particularly in urban districts. Similarly,  
477 larger averages are found in the far north within Al-Burmai and parts of North Al-  
478 Batnah governorates. The large averages were spread sparsely over the eastern coasts  
479 and a clear Eastern excess was apparent along with Jaalan Bani Bu Ali, Al-Duqum ,  
480 Al-Jazir, and rates were high in Sadah and Maribat. Most of the migrants living in  
481 regions such as Al-Duqum are low-skilled migrants (Mansour, 2017) and lack  
482 affordable housing and thus they live in crowded conditions and in close proximity  
483 where protective measures and social distancing are difficult to enforce. Accordingly,  
484 workers who live in such circumstances are at high risk of COVID-19 infection. The

485 low rates were observed mainly in the interior Wilayats of North Al-Batnah (e.g Bidiya,  
486 Al-Qabal, Wadi Bani Khalid, and Jaalan Bani Bu Hassan). These places are  
487 characterised as mountainous environments with low percentages of migrants who  
488 work mostly in the agricultural sector or in domestic work. The spatial distribution of  
489 high and low incidence areas was less heterogeneous in rural and desert areas than  
490 coastal areas.

491 A bivariate map shows the COVID-19 rates among migrants and locals (Omani  
492 populations) across subnational administrative zones (Figure 3). There were marked  
493 spatial differences in the incidence rates between migrants and locals across the study  
494 area. The lowest rates among migrants were seen in the middle part of Al-Dakhalia  
495 governorate (e.g., Bahle wilayat) and a strip from the northeast to the south. On the  
496 other hand, the lowest rates among locals occurred mainly in internal Wilayats in the  
497 West of the country. In essence, these Wilayats are described by suburb and rural  
498 residential areas with a low number of migrants workers. The majority of workers work  
499 in small business properties such as groceries, retail stores, shops, and cafes and they  
500 live in single rooms and less crowded housing units.

501 **Figure 2 spatial distribution of bayes rates of COVID-19 incidence among migrants across the**  
502 **study. The dark blue colour indicates areas with higher rates among Omani populations relative**  
503 **to migrants**

504

505 **Figure 3 Bivariate map of COVID-19 incidence among migrants and Omani populations across**  
506 **the study. The dark red colour refers to areas with higher rates of migrants compared to Omani**  
507 **people.**

#### 508 **4.1. Associations of work sectors with COVID-19 incidence rates**

509 Studying spatial transition mechanisms among migrants' communities in Oman  
510 requires analysing the relatedness between disease infection rates and working sectors

511 and workplaces of migrants' populations. Several maps of excess COVID-19 incidence  
512 rates among migrants were created utilizing work sectors as a base while the infection  
513 cases of migrants were used as an event base (Figure 4). Six factors illustrated  
514 statistically significant spatial correlation with COVID-19 incidence rates. Using the  
515 percentage of migrants working in the agricultural sector as a base, higher than expected  
516 rates were found mostly across the northern Wilayats of both Muscat (e.g. Al-Seeb and  
517 Matrah) and North Al-Batnah (e.g. Sohar and Liwa) governorates. Likewise, other  
518 Wilayats that show greater incidence than expected were in the middle part of the  
519 country particularly Al-Wusta governorate as well as the south-western parts of Dhofar  
520 governorate (Sadah, Mirbat, and Taqah) (Figure 4a). The spatial distribution of excess  
521 disease incidence among migrants for health and social work sectors indicates that  
522 mostly the urban Wilayats within Muscat and Al-Batnah governorates experienced  
523 higher than expected incidence as compared to the average national rate (Figure 4b).  
524 For migrants working in the administrative sector a low number of Wilayats (11) had  
525 greater than expected incidence and these were predominately in the Muscat urban  
526 areas, Al-Duqum in Al-Wusta governorate, and Adam in Al-Dakhliya governorate  
527 (Figure 4c). Considering the retail and business sector as a risk factor determining  
528 COVID-19 incidence, most of the Wilayats located in the northwest showed higher than  
529 expected averages compared to the national average (Figure 4d). Similarly, some  
530 Wilayats in the southeast (e.g. Al-Duqum, Al Jazir, and Mirbat) had higher than average  
531 incidence rates. A cluster of zones extends from the northeast to south (Al-Qabil,  
532 Bidiyah, and Mahawt) associated with below average incidence rates.  
533 Migrants working in the manufacturing sector were associated with increased excess  
534 risks in all administrative zones of Muscat governorate and along Wilayats in the Al-  
535 Batnah coastal plain (Figure 4e). Likewise, a few places along the east and south coasts

536 showed also large incidence rates such as Al-Duqum, Al Jazir, and Salalah. On the other  
 537 hand, a clear belt from north to south (Al-Qabil to Mahawt) was apparent along the  
 538 northeast part of the study whereas this cluster mirrors low incidence rates. For migrants  
 539 working in the mining sector, a few Wilayats experienced high incidence rates of the  
 540 disease particularly Al-Seeb in the north and Salalah in the south (Figure 4f). A cluster  
 541 of low incidence rates can be seen in the northeast part of the Country and most  
 542 Wilayats had infections rates less than the national average.

543 **Figure 4 Spatial distribution of excess COVID-19 incidence rates among migrants in Oman with**  
 544 **respect to the work sectors factors as base covariate, indicating Wilayats that have infection rate**  
 545 **greater than expected incidence: Agricultural sector (a); health and social sectors (b);**  
 546 **administrative sector (c); retails and business (d); manufacturing (e); mining (f).**

547 **Table 2 Bivariate Moran's *I* of the significant risk factors of migrant workers in job**  
 548 **sectors**

Risk factors	Bivariate Moran's I	Pseudo p-value	z-value
Agricultural activities	-0.0172	0.004	-2.360
Retail & business	0.2917	0.003	3.996
Health sector	0.4144	0.002	5.341
Administrative sector	0.4101	0.002	5.152
Manufacturing	0.4119	0.000	6.116
Mining	0.2612	0.013	3.376

549  
 550 A disease conditional map (Fig. 5) shows that high percentages of workers in retail-  
 551 business and manufacturing together were associated with high COVID-19 incidence  
 552 rates in mostly Wilayats located in the north and northwest (upper right map). Whereas,  
 553 the lower rates of disease infection were in the south. The lowest incidence rates also  
 554 cluster in a strip from northeast to the south. The number of locations with emerging

555 COVID-19 incidence conditioned on retails-business and manufacturing were detected  
556 in the upper left panel where manufacturing was high with low retail & business.

557

558 **Figure 5 COVID-19 incidence conditioned on manufacturing and retail business where Wilayats**  
559 **with relatively high rates due to co-location of the two risk factors were represented using the**  
560 **brown colour while the low rates are in light blue.**

561

### 562 **3.3. Spatial autocorrelation and cluster-outlier detection (bivariate LISA)**

563 The bivariate LISA maps in Figure 6 show spatial patterns of the statistically significant  
564 associations between COVID-19 incidence rates among migrants and a set of risk  
565 factors across the subnational boundaries in Oman. The results illustrate that there was  
566 a hotspot and positive autocorrelation (high-high) between COVID-19 incidence rates  
567 and percentage of migrant workers employed in the agricultural sector specifically in  
568 North Al-Batnah governorate as well as Al-Duqum Wilayat (Figure 6a). Positive  
569 autocorrelation (low-low) clustering was found in the northeast of Dhofar governorate.  
570 Two further clusters associated with workers in agriculture (high-low and low-high)  
571 can be observed. The high-low cluster (Wilayats with high disease incidence rates  
572 surrounded by Wilayats with low percentages of migrants' workers in agriculture) was  
573 formed by Wilayats of Jaalan Bani Bu Hassan in the east and Mahadah and Yanqul in  
574 the east. The low-high cluster (Wilayats that had low disease incidence rates surrounded  
575 by areas with high percentages of workers in agricultural sectors), was found only in  
576 the south and composed of Wilayats Al Mazyunah, Rakhyut, and Salalah.

577 Figure 6b shows the map of the spatial association of clusters between disease incidence  
578 rates and migrants' workers in the retail-business sectors. Three positive or hotspot  
579 spatial clusters (high-high) appeared in different locations. The first cluster was found  
580 in the north region (urban zones within Muscat governorate) including Al-Seeb,

581 Muscat, Bowshar and Matrah. The second cluster was formed by Wilayats of Bahla,  
582 Manah, and Nizwa. Two coldspot clusters (low-low) were found in the northeast (Ibri)  
583 and the south (Al Mazyunah, Rakhyut). A high-low cluster type was found in the  
584 northern region of Al-Dakhaliya governorate (Bidbid and Samail) as well as Muqshin  
585 while a cluster of low-high was found in the central part of the country.

586 The association between disease incidence rates and migrants' workers in the  
587 manufacturing sector is depicted in Figure 6c. Administrative zones of positive spatial  
588 association (high-high) were located in Al-Seeb and Bauwshar (within Muscat  
589 governorate), Sohar in the north, and Al-Duqum in the central east of the study. Cold  
590 spots comprise a spatial cluster of low-low were located in the Mauhuat and Sadah  
591 Wilayats. Figure 6d represents the relationship between incidence rates and migrant  
592 workers in the health sector. Positive autocorrelation pattern (high-high) was observed  
593 in the north, especially in urban zones (Al-Seeb, Bauwshar, Matrah, Barka and Muscat).  
594 The spatial pattern of cold spots (low-low) appears mostly in the far north and  
595 northwest. A few statistically significant (at the 0.05 level) high-low wilayats (outliers)  
596 occurred in the north of Al-Dakhaliya governorate (Bidbid). This pattern demonstrates  
597 that a high prevalence of COVID-19 among migrants coexisted with migrant workers  
598 in the health sector.

599

600 For the association between disease incidence rates and migrants' workers in the  
601 administrative sectors (Figure 6e), our results showed three high-high cluster types in  
602 Al-Seeb and Barka in the north, Adam and Bahla in Al-Dkhaliya governorate, and the  
603 southwest of Dhofar governorate (Sadah and Mirbat). The low-low clusters were found  
604 in the east of Dhofar governorate (Thumrayt and Muqshin). On the other hand,  
605 outlier clusters (high-low and low-high) can be observed in the north (Ar Rustaq, Ibra,



606 and Samail) and the far South (Al Mazyunah and Dalkut), respectively. The high  
607 association between disease incidence rates and migrants' workers is spatially  
608 concentrated in the north (Sohar, Al-Seeb, and Bowsher) and in the east (Al-Duqum )  
609 (Figure 6f). In contrast, a cold spot (low–low) pattern is observed in the northwest part  
610 of the country (Ibri, Bahla, and Adam).

611

612 **Figure 6 Maps of Local indicators of Spatial Association (LISA) bivariate clusters of COVID-19**  
613 **incidence rates among migrants with migrant workers in 6 sectors: Agriculture sector (a) retail-**  
614 **business sectors (b) manufacturing sector (c) health (d) administrative sector (e) mining sector (f).**  
615 **Hotspot (High-High), Coldspot (Low-Low), and outliers (High-Low & Low-High). The statistical**  
616 **inference based on Monte Carlo randomisation test at 999 permutations, showing significant**  
617 **pseudo  $p < 0.05$  clusters.**

618

## 619 **5. Discussion**

620 Although a considerable body of existing literature has addressed the interrelationships  
621 between COVID-19 and migrants (Turcotte & Savage, 2020; Keller and Wagner, 2020;  
622 Wernly et al., 2020; Greenaway et al., 2020; Openshaw & Travassos, 2020), spatial  
623 studies on the associations between migrants work sectors as risk factors, and disease  
624 incidence rates have not been conducted. Accordingly, this research represents a novel  
625 contribution to the evidence base concerning associations between employment sectors  
626 of migrants and COVID-19 incidence. It also offers useful insights regarding the spatial  
627 autocorrelation and spillover of the disease among migrants in Oman. Furthermore, it  
628 represents a novel linkage of a wide range of health, socioeconomic and  
629 epidemiological fields. Although migrants, especially low skilled workers, play an  
630 essential role in the response to the COVID-19 pandemic through working on the

631 frontlines, they may be some of the most affected and most vulnerable populations to  
632 disease infection (Claudia, 2020).

633 The findings from the analyses showed that six main variables (migrants working in  
634 employment sectors) were strongly associated with spatial variations in excess COVID-  
635 19 incidence across Oman. The spatial clustering of high COVID-19 incidence and  
636 their relationship with work sectors is clear in urban Wilayats within Muscat  
637 governorate as well as Al-Batnah coastal plain. This can be attributed to the  
638 demographic characteristics of these areas where large number of migrants are  
639 concentrated and working in several public and private sectors such as commercial  
640 properties, agricultural farms, shops, shopping centres, warehouses, factories, medical  
641 centres and hospitals.

642

643 In this research, the main focus was on the recognition of the workplace as a COVID-  
644 19 infection risk. We found a number of risk factors with significant associations with  
645 COVID-19 incidence rates in Oman. The significant autocorrelation between these  
646 factors and disease incidence among migrants indicates that these factors might be  
647 indispensable predictors of infection occurrence at the subnational level. Considering  
648 the percentages of migrants working in agriculture, high rates of COVID-19 incidence  
649 were found in the north of the country particularly along the Al-Batnah plain. This  
650 pattern might be explained by the fact that the low wage farmworkers in this sector face  
651 challenges to afford the required prevention and measures such as isolation, social  
652 distancing and disinfectant. In addition, low skilled workers in rural and desert areas  
653 often have relatively poor knowledge about the disease prevention with low access to  
654 health facilities. Consequently, high percentages of migrants working in agricultural  
655 activities and operations were associated with greater levels of COVID-19 incidence

656 rates across local communities. Our findings indicated that the prevalence of COVID-  
657 19 among health and medical workers was significantly higher than for others sectors  
658 which is consistent with previous studies, (Kasper,2020; Godderis, 2020). Accordingly,  
659 jobs in the health and medical sectors can be classified as presenting high or very high  
660 exposure risks to COVID-19 infection. Most excess disease incidence among migrants  
661 working in health and medical sectors were in urban zones within Muscat and Al-  
662 Batnah in the north and Al-Duqum in the east. The high stress of patient loads in  
663 hospitals and clinics makes workers in these sectors particularly vulnerable to infection.  
664 LISA analysis revealed that the bivariate clustered zones with High-High positive  
665 spatial association for COVID-19 incidence and migrant workers in the health sector  
666 were located around the capital Muscat, as well as Barka Wilayat in South Al-Batnah.  
667 Local clusters of negative spatial association Low-High in the south and southeast.  
668 Cluster detection also identified four clusters exhibiting positive autocorrelation  
669 between disease incidence and workers in the manufacturing sector in the north, within  
670 Muscat governorate and in the northwest around Sohar Wilayat, and Al-Duqum in the  
671 east. In addition, spatial association was found between COVID-19 incidence rates  
672 among migrants and percentages of migrants working in administrative sectors.  
673 Positive associations (High-High) were found mainly in the urban areas of Muscat  
674 governorate. The high Covid-19 incidence rates in this cluster can be explained by the  
675 fact that large number of migrants are working in public and private establishments and  
676 organizations within Muscat governorates and the majority live in overcrowded  
677 housing. We identified three hotspot (high-high) clusters of the association between  
678 workers in the retail and business operations and disease rates. These clusters included  
679 10 Wilayats located in the north and south of the study area with incidence rates higher  
680 than the national average surrounded by other zones with higher than average rates.

681 This can be attributed to the fact that most retail stores, including drug stores, grocery  
682 stores, and other entities that sell essential supplies, and businesses in the informal  
683 sector, remained open during the COVID-19 pandemic, particularly those with high  
684 customer volumes.

685 This research could provide effective information to decision makers by identifying  
686 disease infection hotspots that are associated with workplace relative risks.  
687 Consequently, precise measures, prevention and control of the disease can be  
688 implemented. The ramifications of lockdown and social distance restrictions will be felt  
689 for months, if not years, particularly for migrant workers within their work systems.  
690 Despite appearing to be at the first stage of disease receding and vaccination  
691 programmes beginning, migrant workers remain the most vulnerable groups to  
692 infection.

693 The spatial aspect of COVID-19 associations with workplaces has not been studied, but  
694 it can provide vital information for the effective interventions that target protecting  
695 workers from disease infections in local areas. In the absence of individual-level data  
696 of COVID-19 infection, particularly across the developing nations, aggregated area-  
697 level variables are often the only source of information to investigate how factors  
698 associated with disease incidence may contribute to understanding infection patterns  
699 and risk mitigation. Workplace sectors of migrants have been identified as potential  
700 determinants of COVID-19 incidence rates among migrants at the subnational spatial  
701 scale. As a result, policy makers and governmental planners need to address  
702 geographically the associations between disease transmission and workers in each  
703 economic sector ,particularly with the lifting or loosening of lockdown restrictions. To  
704 reduce infection risk exposure within workplaces, the implementation of administrative  
705 control is required which includes several procedures and measures such as excluding

706 sick workers and visitors, appropriate social distance and cleaning and disinfection  
707 policies.

708 Data at the household or individual-level were not available, and ,therefore, this study  
709 was limited by the utilized aggregated dataset at the zonal subnational level.  
710 Accordingly, the explanations of the results should be interpreted with caution to avoid  
711 the risk of ecological fallacies. In addition, we included only the risk factors that are  
712 associated with job sectors and workplaces while several risk variables that might be  
713 interrelated to COVID-19 incidence were not considered due to data availability.

714

715 An understanding of the associations between the disease incidence and migrant  
716 workers is indispensable to control the spread of the virus, specifically among workers  
717 within workplace environments. The results of this study showed that migrants working  
718 in the health care system in the capital city of Muscat are infected by the disease at a  
719 high rate, while more cases were found among migrants working in the agricultural  
720 sector in Al-Batinah area. This reflects the economic characteristics of these two areas  
721 and how migrant workers in these areas are impacted by working and living  
722 circumstances. Furthermore, the spatial distribution of migrants based on workplaces  
723 resulted in differences in the spatial distribution of incidence between migrants and  
724 Omani citizens across the country. Migrants in the country have the right to access the  
725 health system similarly to Omani citizens; however, the type of work that migrants  
726 occupy and the associated social characteristics maybe the main determinant of the  
727 increased number of cases amongst migrants. Decision- makers may need to implement  
728 different mitigation measures where migrants are clustered to slow down the spread of  
729 the disease.

730

731 **6. Conclusion**

732 Oman is major importers of migrant workers in the region and globally. The dominance  
733 of migrants is more pronounced in the workforce than it is in the total population and  
734 hence migrants are major contributors to the economic prosperity of the country.  
735 Migrants in Oman work in many economic sectors and are distributed in different areas  
736 of the country. The socioeconomic conditions of migrants in Oman and the region make  
737 them more vulnerable and at high risk of suffering from the COVID-19 pandemic.  
738 Therefore, the aim of this study was to model the spatial distribution of the COVID-19  
739 among migrants based on the type of workplace to understand the spatial relationships  
740 between migrant infection rates of COVID-19 and their workplace.

741 Our analysis emphasizes the associations between percentages of migrants in their  
742 workplaces and disease incidence at subnational and finer scale. The increased  
743 percentages of migrant workers in the agricultural, health and business sectors were  
744 associated with an excess of COVID-19 incidences mainly in Muscat and Al-Batnah  
745 governorates. Similarly, migrant workers in manufacturing and mining were associated  
746 with increased risk of COVID-19 incidence rates in the east particularly (Al-Duqm)  
747 and the south (Salalah). Investigating the socioeconomic, cultural and systemic factors  
748 that may lead to the spread of COVID-19 among migrants is important for developing  
749 and monitoring targeted preventive and intervention strategies for reduce transmission  
750 of the disease. The results of this study can aid in understanding the dynamics of the  
751 disease among migrants and hence the processes controlling its spread among migrants  
752 over space and time. This can contribute in helping decision-makers to adopt more  
753 appropriate mitigation measures and actions to control the propagation of the disease  
754 in Oman and the other GCC countries.

755

756 **References**

757 Abulibdeh, A. (2020). Can COVID 19 mitigation measures promote telework  
758 practices? *Journal of Labor and Society*, 23(4). Pp: 551–576.

759

760 Al-Kindi, K., Alkharusi, A., Alshukaili, D., Al Nasiri, N., Talal Al-Awadhi, T.,  
761 Charabi, Y., and El Kenawy, A. M. (2020). Spatiotemporal Assessment of COVID-19  
762 Spread over Oman Using GIS Techniques. *Earth Systems and Environment* (2020)  
763 4:797–811. <https://doi.org/10.1007/s41748-020-00194-2>

764

765 Al-Maadid, A., Guglielmo Maria Caporale, G. M., Spagnolo, F., Spagnolo, N. (2020).  
766 The impact of business and political news on the GCC stock markets. *Research in*  
767 *International Business and Finance* 52 (2020) 101102.

768

769 Alzamora MC, Paredes T, Caceres D, Webb CM, Valdez LM, La Rosa M (2020) Severe  
770 COVID-19 during pregnancy and possible vertical transmission. *Am J Perinatol*  
771 37:861. <https://doi.org/10.1055/s-0040-1710050>

772

773 Bahrini, R., and Filfilan, A. (2020). Impact of the novel coronavirus on stock market  
774 returns: evidence from GCC countries. *Quantitative Finance and Economics*. Volume  
775 4, Issue 4, 640–652.

776

777 Borjas, G. J., and Cassidy, H. (2020). The Adverse Effect of the COVID-19 Labor  
778 Market Shock on Immigrant Employment. IZA DP No. 13277

779

780 Bourgonje AR, Abdulle AE, Timens W, Hillebrands JL, Navis GJ, Gordijn SJ, Bolling  
781 MC, Dijkstra G, Voors AA, Osterhaus AD (2020) Angiotensin-converting enzyme-2  
782 (ACE2), SARS-CoV-2 and pathophysiology of coronavirus disease 2019 (COVID-19).  
783 *J Pathol*. <https://doi.org/10.1002/path.5471>

784 **Budnitz, H., & Tranos, E. (2021). Working from home and digital divides: resilience**  
785 **during the pandemic. *Annals of the American Association of Geographers*, 1-21.**

786

787 Chakraborty I, Maity P (2020) COVID-19 outbreak: Migration, effects on society,  
788 global environment and prevention. *Sci Total Environ*. <https://doi.org/10.1016/j.scitotenv.2020.138882>

790

791 Clark, E., Fredricks, K., Woc-Colburn, L., Bottazzi, M. E., and Weatherhead, J. (2020).  
792 Disproportionate impact of the COVID-19 pandemic on immigrant communities in the  
793 United States. *PLoS Negl Trop Dis* 14(7): e0008484.  
794 <https://doi.org/10.1371/journal.pntd.0008484>

795

796 Costa, R., and de Valk, H.A.G. (2018). Ethnic and socioeconomic segregation in  
797 Belgium: a multiscalar approach using individualized neighbourhoods. *European*  
798 *Journal of Population* 34(2): 225–250. <https://doi.org/10.1007/s10680-018-9480-6>

799

800 Das, K. C., & Gokhale, N.. (2009). Localization of labor and international migration:  
801 A case study of the sultanate of Oman. Paper presented at the XXVI International  
802 Population Conference of the International Union for the Scientific Study of  
803 Population.

804

805 De Bel-Air, F. (2015). Demography, migration, and the labour market in  
806 Oman, GLMM- EN -No. 9/2015. Migration Policy Centre: European University  
807 Institute.

808

809 Elmousalami HH, and Hassanien AE. 2020. Day level forecasting for Coronavirus  
810 Disease (COVID-19) spread: analysis, modeling and recommendations.  
811 [arxiv.org/ftp/arxiv/papers/2003/2003](https://arxiv.org/ftp/arxiv/papers/2003/2003)

812

813 European Centre for Disease Prevention and Control. Guidance on Infection and  
814 Prevention Control of Coronavirus Disease (COVID-19) in Migrant and Refugee  
815 Reception and Detention Centres in the EU/EEA and the United Kingdom Stockholm.  
816 ECDC, 2020.

817

818 Fernandez, B. (2010). Cheap and disposable? the impact of the global economic crisis  
819 on the migration of Ethiopian women domestic workers to the Gulf. *Gender and*  
820 *Development*, 18(2).

821



822 Fernandez, B. (2011). Household help? Ethiopian women domestic workers' labor  
823 migration to the Gulf countries. *Asian and Pacific Migration Journal*, 20(3–4), 433–  
824 457.

825

826 Greenaway, C., Hargreaves, S., Barkati, S., Coyle, C. M., Gobbi, F., Veizis, A., and  
827 Douglas, P. (2020). COVID-19: Exposing and addressing health disparities among  
828 ethnic minorities and migrants. *Journal of Travel Medicine*, 2020, 1–3. doi:  
829 10.1093/jtm/taaa113

830

831 Greenaway, C., and Castelli, F. (2019). Infectious diseases at different stages of  
832 migration: an expert review. *J Travel Med* 2019; 26: taz007. doi: 10.1093/jtm/taz007.

833 Gupta A, Pradhan B, Maulud KNA (2020) Estimating the Impact of Daily Weather on  
834 the Temporal Pattern of COVID-19 Outbreak in India. *Earth Syst Environ*. <https://doi.org/10.1007/s41748-020-00179-1>

835

836  
837 Hardy LJ, Getrich CM, Quezada JC, Guay A, Michalowski RJ, Henley E. A call for  
838 further research on the impact of state-level immigration policies on public health. *Am*  
839 *J Public Health*. 2012; 102(7):1250–4. <https://doi.org/10.2105/AJPH.2011.300541>  
840 PMID: 22594736.

841

842 Hsu A, Lane N, Sinha S et al (2020). Understanding the Impact of COVID-19 on  
843 Residents of Canada's Long-Term Care Homes – Ongoing Challenges and Policy  
844 Responses. International Long Term Care Policy Network, 2020.

845

846 Huang, X., Lu, J., Gao, S., Wang, S., Liu, Z., & Wei, H. (2021). Staying at Home Is a  
847 Privilege: Evidence from Fine-Grained Mobile Phone Location Data in the United  
848 States during the COVID-19 Pandemic. *Annals of the American Association of*  
849 *Geographers*, 1-20.

850

851 Kar, A., Le, H. T., & Miller, H. J. (2021). What Is Essential Travel? Socioeconomic  
852 Differences in Travel Demand in Columbus, Ohio, during the COVID-19 Lockdown.  
853 *Annals of the American Association of Geographers*, 1-24.

854

855 Keller, A. S., and Wagner, B. D. (2020). COVID-19 and immigration detention in the  
856 USA: time to act. *www.thelancet.com/public-health* Vol 5 May 2020.  
857

858 Khunti, K., Singh, A. K., Pareek, M., and Hanif, W. (2020). Is ethnicity linked to  
859 incidence or outcomes of covid-19? *BMJ* 2020; 369:m1548.  
860

861 Li W, Thomas R, El-Askary H, Piechota T, Struppa D, Ghaffar KAA (2020)  
862 Investigating the significance of aerosols in determining the coronavirus fatality rate  
863 among three European Countries. *Earth Syst Environ.* [https://doi.org/10.1007/s41748-](https://doi.org/10.1007/s41748-020-00176-4)  
864 [020-00176 -4](https://doi.org/10.1007/s41748-020-00176-4)  
865

866 Liu J, Zhou J, Yao J, Zhang X, Li L, Xu X, He X, Wang B, Fu S, Niu T (2020a) Impact  
867 of meteorological factors on the COVID-19 transmission: a multi-city study in China.  
868 *Sci Total Environ.* <https://doi.org/10.1016/j.scitotenv.2020.138513>  
869

870 Mansour, S., Al Kindi, A., Al-Said, A., Al-Said, A., and Atkinson, P. (2021).  
871 Sociodemographic determinants of COVID-19 incidence rates in Oman: Geospatial  
872 modelling using multiscale geographically weighted regression (MGWR). *Sustainable*  
873 *Cities and Society* 65 (2021) 102627  
874

875 Mansour, S. (2017). Spatial concentration patterns of South Asian low-skilled  
876 immigrants in Oman: A spatial analysis of residential geographies. *Applied Geography*  
877 88 (2017) 118–129  
878

879 Maroko, A.R., Nash, D., and Pavilonis, B.T. (2020). COVID-19 and inequity: a  
880 comparative spatial analysis of new York City and Chicago hot spots. *J Urban Health*  
881 97(4):461–470. <https://doi.org/10.1007/s11524-020-00468-0>  
882

883 Martinez O, Wu E, Sandfort T, Dodge B, Carballo-Diequez A, Pinto R, et al. Evaluating  
884 the impact of immigration policies on health status among undocumented immigrants:  
885 a systematic review. *J Immigr Minor Health.* 2015; 17(3):947–70.  
886 <https://doi.org/10.1007/s10903-013-9968-4> PMID: 24375382  
887

888 Nicola M, Alsafi Z, Sohrabi C, Kerwan A, Al-Jabir A, Iosifidis C, Agha M, Agha R  
889 (2020b) The socio-economic implications of the coronavirus and COVID-19 pandemic:  
890 a review. *Int J Surg*. <https://doi.org/10.1016/j.ijisu.2020.04.018>.  
891

892 Oyedotun, T. D. T., & Moonsammy, S. (2021). Spatiotemporal variation of COVID-19  
893 and its spread in South America: A rapid assessment. *Annals of the American*  
894 *Association of Geographers*, 111(6), 1868-1879.  
895

896 Openshaw, J. J., and Travassos, M. A. (2020). COVID-19 outbreaks in U.S. immigrant  
897 detention centers: the urgent need to adopt CDC guidelines for prevention and  
898 evaluation. *Clin Infect Dis* 2020; ciaa692, <https://doi.org/10.1093/cid/ciaa692>.  
899

900 Platt, L., and Warwick, R. (2020). Are some ethnic groups more vulnerable to COVID-  
901 19 than others? 2020; 1–27. <https://www.ifs.org.uk/inequality/chapter/are-some-ethnic-groups-more-vulnerable-to-covid-19-than-others/>.  
902  
903

904 Shah, N. M. (2004a). Arab migration patterns in the Gulf. *IOM* (2004), 91–113.  
905

906 Shah, N. M. (2004b). Gender and labour migration to the Gulf countries. *Feminist*  
907 *Review*, 77, 183–185.  
908

909 Shereen MA, Khan S, Kazmi A, Bashir N, Siddique R (2020) COVID-19 infection:  
910 origin, transmission, and characteristics of human coronaviruses. *J Adv Res*. <https://doi.org/10.1016/j.jare.2020.03.005>  
911  
912

913 Sohrabi C, Alsafi Z, O’Neill N, Khan M, Kerwan A, Al-Jabir A, Iosifidis C, Agha R  
914 (2020) World Health Organization declares global emergency: a review of the 2019  
915 novel coronavirus (COVID-19). *Int J Surg*. <https://doi.org/10.1016/j.ijantimicag.2020.105948>  
916

917 Sönmez, S., Apostolopoulos, Y., Lemkeb, M. K., and Hsieh, J. (2020). Understanding  
918 the effects of COVID-19 on the health and safety of immigrant hospitality workers in  
919 the United States. *Tourism Management Perspectives* 35, July 2020, 100717  
920

921 Tai, D.B.G., Doubeni S. A. C. A, IG, S., Wieland, M. L. (2020). The disproportionate  
922 impact of COVID-19 on racial and ethnic minorities in the United States. *Clin Infect*  
923 *Dis* 2020; ciaa815. doi: 10.1093/cid/ciaa815.

924

925 Turcotte, M., and Savage, K. (2020). The Contribution of Immigrants and Population  
926 Groups Designated as Visible Minorities to Nurse Aide, Orderly and Patient Service  
927 Associate Occupations. Ottawa, Canada: Statistics Canada, 2020.

928

929 Wernly, B, Wernly, S, Magnano, A., and Paul, E. (2020). Cardiovascular health care  
930 and health literacy among immigrants in Europe: a review of challenges and  
931 opportunities during the COVID-19 pandemic. *Journal of Public Health: From Theory*  
932 *to Practice*. <https://doi.org/10.1007/s10389-020-01405-w>.

933

934 Willoughby, J. (2006). 13 ambivalent anxieties of the south Asian–gulf Arab labor  
935 exchange. *Globalization and the Gulf*. 223.

936

937 World Bank (2021).

938 <https://data.worldbank.org/indicator/SP.POP.TOTL?locations=OM>. (Access,  
939 January, 2021)

940

941 Worldometer (2021). <https://www.worldometers.info/coronavirus/country/oman/>  
942 (Access on 29/1/2021)

943

944 *Zhai, W., Liu, M., Fu, X., & Peng, Z.-R. (2021). American Inequality Meets COVID-*  
945 *19: Uneven Spread of the Disease across Communities. Annals of the American*  
946 *Association of Geographers, 1-21.*

947

948 **SHAWKY MANSOUR** is an Associate Professor of GIS in the Department of  
949 Geography and GIS, Faculty of Arts, Alexandria University, Egypt, and in the  
950 Geography Department, College of Arts and Social Sciences, Sultan Qaboos  
951 University, Muscat, Oman. E-mail: [shmansour@squ.edu.om](mailto:shmansour@squ.edu.om). He is a specialist in GIS  
952 with particular interests in GIScience and spatial modeling. His research focuses on  
953 developing and utilizing advanced geospatial techniques to model and analyze the  
954 interrelationships between socioeconomic, demographic and environmental  
955 phenomena.

956

957 **AMMAR ABULIBDEH** is an Assistant Professor in the Department of Humanities,  
958 College of Arts and Science, Qatar University, Doha, Qatar. E-mail:  
959 [aabulibdeh@qu.edu.qa](mailto:aabulibdeh@qu.edu.qa). His research focuses on smart urban planning and design,  
960 sustainable built environment, sustainable transportation, and the water–energy–food  
961 nexus.

962

963 **MOHAMMED ALAHMADI** is an Associate Professor in the National Center for  
964 Remote Sensing Technology at King Abdulaziz City for Science and Technology  
965 (KACST), Riyadh, Saudi Arabia. E-mail: [mhalahmadi@kacst.edu.sa](mailto:mhalahmadi@kacst.edu.sa). He is an expert in  
966 modelling small area population data from satellite data. His research focuses on the  
967 application of machine learning, space-time modelling and global environmental  
968 change.

969 **ADHAM AL-SAID** is an Assistant Professor in the department of Economics and  
970 Finance at Sultan Qaboos University, Muscat, Oman, Email: [adham@squ.edu.om](mailto:adham@squ.edu.om). His  
971 research interests include the macroeconomic Policy, Economic Development and  
972 Economic Diversification.

973 **ALKHATTAB AL-SAID** is a part-time faculty member, College of Economics and  
974 Political Sciences, Sultan Qaboos University, Muscat, Oman, Email:  
975 [alkhattabalsaid@gmail.com](mailto:alkhattabalsaid@gmail.com). His research interests include the political economy of the  
976 middle east and Gulf studies.

977 **GARY WATMOUGH** is an Interdisciplinary Lecturer in Landuse and socioecological  
978 systems in the School of Geosciences, University of Edinburgh, Scotland. email:  
979 [gary.watmough@ed.ac.uk](mailto:gary.watmough@ed.ac.uk). Gary's main research focus is in the development of  
980 approaches for geographical targeting of resources particularly in low- and middle-  
981 income countries. Gary's research into Earth Observations for Sustainable  
982 Development Goals (EO4SDGs) supports the UN's call for a data revolution to help  
983 monitoring progress towards socioeconomic indicators.

984 **PETER M. ATKINSON** is a Distinguished Professor of Spatial Data Science and  
985 Executive Dean of the Faculty of Science and Technology at Lancaster University,  
986 Bailrigg, Lancaster LA1 4YW, UK. E-mail: [pma@lancaster.ac.uk](mailto:pma@lancaster.ac.uk). He is currently a  
987 Visiting Professor at the University of Southampton, Southampton, UK and the Chinese  
988 Academy of Sciences, Beijing, China. His research agenda focuses on the development  
989 and application of spatial statistical and data science methods, coupled with remote  
990 sensing, to tackle some of the most important challenges in environment and  
991 epidemiology facing humankind.

992